Data Munging Project Description Your team has been tasked with collecting metrics on a plethora of disparate shipping data. This task comes straight from the top, so it would be wise to give it your all. The data is contained in a number of different spreadsheets, each with its own competing schema. In order to interrogate the data, all of it has to be in the same place and in the same format. Currently, the shipping data exists in several places in several formats and is therefore impossible to query. To combine the spreadsheets, you need to write a python script to read through every row, extract the pertinent data, figure out how to combine it, munge it into the right format, and upload it to the database. Plenty of steps, but the resulting data will be much easier to query. Once the database contains all the data, you can pass it off to the analysis team to extract all the relevant metrics. Good luck! **Data Dictionary** Part 1: Get the data First, you need to get your hands on the relevant data. The shipping department has been kind enough to provide you with a repository containing all of their spreadsheets, as well as a copy of the sqlite database. First, fork and clone the repository at: https://github.com/theforage/forage-walmart-task-4 Part 2: Populate the database Your task is to insert all of the data contained in the provided spreadsheets into the SQLite database. You will write a Python script which: • Reads each row from the spreadsheets. Extracts the relevant data. Munges it into a format that fits the database schema. • Inserts the data into the database. Spreadsheet 0 is self contained and can simply be inserted into the database, but spreadsheets 1 and 2 are dependent on one another. Spreadsheet 1 contains a single product per row, you will need to combine each row based on its shipping identifier, determine the quantity of goods in the shipment, and add a new row to the database for each product in the shipment. The origin and destination for each shipment in spreadsheet 1 are contained in spreadsheet 2. You may assume that all the given data is valid - product names are always spelled the same way, quantities are positive, etc. When you're finished, convert the python script you used to populate the database into a PDF and submit it below. **Import Libraries** import pandas as pd import numpy as np import matplotlib.pyplot as plt import seaborn as sns import random import datetime from datetime import datetime, timedelta import scipy.stats #import sqlite3 import sqlite3 as sq3 import warnings warnings.filterwarnings('ignore') %matplotlib inline #sets the default autosave frequency in seconds **%autosave** 60 sns.set style('dark') sns.set(font scale=1.2) plt.rc('axes', titlesize=9) plt.rc('axes', labelsize=14) plt.rc('xtick', labelsize=12) plt.rc('ytick', labelsize=12) pd.set option('display.max columns', None) #pd.set option('display.max rows',None) pd.set option('display.width', 1000) pd.option context('float format','{:.2f}'.format) random.seed(0) np.random.seed(0) np.set printoptions(suppress=True) Autosaving every 60 seconds %%capture %load_ext sql #%sql sqlite://chinook.db $\#\$sql\ mysql://studentuser:studentpw@localhost/dognitiondb$ #%sql USE dognitiondb **Load CSV files** df0 = pd.read csv("shipping data 0.csv") origin_warehouse destination store product on_time product_quantity driver_identifier d5566b15-b071-4acf-8e8e-50d33715-4c77-4dd9-8b9dd8da0460-cf39-4f38-9fff-0 lotion 59 c98433083b2d ff1ca372a2a2 6c9b4e344d8a c42f0de8-b4f0-4167-abd1-172eb8f3-1033-4fb6-b66b-293ccaec-6592-4f04-aae5windows True 28 ae79e5e18eea d0df09df3161 3e238fe62614 b145f396-de9b-42f1-9cc9-65e4544d-42ae-4751-9580-80988f09-91a3-4e1b-8e69-2 skis True 63 f5b52c3a941c bdcb90e5fcda 13551c53f318 f4372224-759f-43b3-bc83-745bee4e-710c-4538-8df1-5f79b402-655f-4d8e-8ff3-3 bikes 47 ca6106bba1af 5c146e1092a6 5ef05870e0ad 49d0edae-9091-41bb-a08d-425b7a1a-b744-4c6b-898e-58beb5d3-98f8-4077-a964-4 False 73 candy ab1c66bd08d5 d424dd8cf18e 1f04f7cb11e5 d2ee1b75-2218-4753-9487-0a994581-341f-43bf-979da9784b8d-d222-4cdf-93fb-105 paint True 95 dcca23d667c6 ece1e58de7ec b3886c8033c5 6a6d3fce-c5aa-4154-a6a3-403bf915-a897-4918-933b-2fd9a976-bac5-4803-be43-106 False 54 snakes b56cb41f709f 3996e144e960 bf93cc618ad1 b19cec0d-357e-4c6b-9257d3b17672-60fb-443f-a047-45c9bd5b-caf6-4ec1-b1eb-107 alternators False 20 8be52b1c71b5 2c379132dcb1 09fe615fbdc6 d2a2460e-00d1-41f2-84ccb9f78d5b-79ae-441e-9dbfd7432792-20ad-4a7f-a395pencil 7 108 False eba01eb88d75 592767af34a5 sharpeners 81f04fee89fe 75891066-59b4-437b-951f-28fff0d2-38ea-40a7-b2efcebc86e8-c327-46f7-96b3-109 False 35 apples ec553fb26b94 c2a2f7e69370 35684d169455 110 rows × 6 columns In [4]: df0.columns Index(['origin warehouse', 'destination store', 'product', 'on time', 'product quantity', 'driver identifier'], Out[4]: dtype='object') df0.drop(['origin_warehouse', 'destination_store','on_time', 'product_quantity'], axis=1, inplace=True) df0.head() product driver_identifier lotion d8da0460-cf39-4f38-9fff-6c9b4e344d8a windows 293ccaec-6592-4f04-aae5-3e238fe62614 2 80988f09-91a3-4e1b-8e69-13551c53f318 3 5f79b402-655f-4d8e-8ff3-5ef05870e0ad 58beb5d3-98f8-4077-a964-1f04f7cb11e5 df0.columns = ["name", "id"] df0.head() name d8da0460-cf39-4f38-9fff-6c9b4e344d8a lotion 293ccaec-6592-4f04-aae5-3e238fe62614 windows 2 skis 80988f09-91a3-4e1b-8e69-13551c53f318 3 bikes 5f79b402-655f-4d8e-8ff3-5ef05870e0ad 58beb5d3-98f8-4077-a964-1f04f7cb11e5 df0_new = df0[["id","name"]] df0 new id name d8da0460-cf39-4f38-9fff-6c9b4e344d8a lotion 293ccaec-6592-4f04-aae5-3e238fe62614 windows 2 80988f09-91a3-4e1b-8e69-13551c53f318 skis 5f79b402-655f-4d8e-8ff3-5ef05870e0ad bikes 58beb5d3-98f8-4077-a964-1f04f7cb11e5 candy a9784b8d-d222-4cdf-93fb-b3886c8033c5 paint 106 2fd9a976-bac5-4803-be43-bf93cc618ad1 45c9bd5b-caf6-4ec1-b1eb-09fe615fbdc6 alternators 108 d7432792-20ad-4a7f-a395-81f04fee89fe pencil sharpeners cebc86e8-c327-46f7-96b3-35684d169455 apples 110 rows × 2 columns #df0 new.to csv("product.csv",index=False) df1 = pd.read_csv("shipping_data_1.csv") df1 shipment_identifier product on_time **0** 449263b4-6c93-4f19-8b6a-0d99a29fc637 False pants 449263b4-6c93-4f19-8b6a-0d99a29fc637 pants False 449263b4-6c93-4f19-8b6a-0d99a29fc637 False pants 449263b4-6c93-4f19-8b6a-0d99a29fc637 False keyboards 449263b4-6c93-4f19-8b6a-0d99a29fc637 False keyboards 105 c2237ca1-b7e3-40ab-b798-e1ea469301dc keyboards True 106 cfa8a834-54bd-4f47-99ca-8912df32913b animal masks False furniture 107 cfa8a834-54bd-4f47-99ca-8912df32913b False 108 cfa8a834-54bd-4f47-99ca-8912df32913b furniture False 109 cfa8a834-54bd-4f47-99ca-8912df32913b furniture False 110 rows × 3 columns df2 = pd.read_csv("shipping_data_2.csv") df2 shipment_identifier driver_identifier origin_warehouse destination_store 449263b4-6c93-4f19-8b6abb75bf7d-c008-4267-bf92-5e9405de-a078-4b00-99c6c12025e6-6f9c-4728-8c3c-0 0d99a29fc637 6089cff5fe56 96564568b63c 9f840bde6f1a 76e5b84a-9d09-4efb-8b43-372fd2b1-b2a7-4553-b6d7e34973c8-9ca9-4a06-b497-85b8d394-a67c-48b6-b1de-1 55be323ba622 a0c932b958bb 7a8b49625fc2 426a1bc88e56 b541a47d-89b1-4805-97d0-469d957f-28ef-4eac-956afcadc756-61e9-41bb-871b-47bdfc40-f3db-4678-b6a7-2 1988832321f1 d2a42b06d3ab 43f1e1c2fd32 d3546c5aa981 3fc6b63d-27b4-408c-b3b3cd140190-a53b-4660-a5b4-89ba200c-ca90-443a-b64f-5ae3e541-2098-45b6-8d94-3 397bce091eae e8e94a45b079 cc844a6506f0 35d176185606 491ee4e8-be80-4f52-802b-7aebe820-8478-4a29-a606c6addf8b-eea6-43b8-9040-1f228b52-7165-4d7f-a731-4 d8fe1a6bd487 b5620b1a0d99 7c59af677e24 3f7707aefb2a 22768e96-0dad-40d2-8204d57d76d8-7dca-4ee4-84c0-5d64f731-cb01-4992-a27c-9136e027-dc50-48b0-b2cf-5 d01f0c412815 3921263c3826 a6e1342f4913 1745fb4c8779 f20bbd93-1312-4f70-b257abc09fec-2fa0-48f6-b7c4-52479603-9957-4e4b-91eb-85f31f19-81ff-4b03-b862-6 654056412ec5 913620785520 337c358d1755 f2ba16605434 bb53f18b-e3c5-48c1-900e-192cc6dc-4799-4247-a228-1add84b7-14f5-4857-903bab78787d-6f8d-48ec-8cef-7 6d198675c00b 578408246946 e8ed623ca467 a7a50ae9c7db ee67c3b0-aa89-4b3b-8bbc-4159e22a-d107-42e6-ba56e31e22c1-5395-43d8-8a0afa0ce0bb-b0d8-469d-8d42-8 e1153fc48272 79396d627f66 9d70695c132b f9b65ad8df08 f04f3daf-8ede-4787-a3ad-6060c04b-921b-4bf9-b2d3-130208de-fef4-46cd-8b9bd194a942-695a-4e09-9701-9 6ff06274229d 1ea5b939895b 40e57257e5c6 490ea8d627f6 d2306016-fe82-41f8-a8e1dc042557-cbee-4743-9b72-48c4ca28-4db8-420e-af57-469c402e-d073-49a4-9598-10 b06812007036 241818a81194 2c0a34a99cc2 c44a32724643 3433af6d-4857-4dfa-88d6-57001f3e-d6be-4031-9295-49714439-c858-4e61-b7c6-0966463d-fbac-41cb-b780-11 0211b98c9beb 5dcd2a2fce6b a6b208c0ad46 f4d4c848c46b c42221be-4851-42df-9184-474067dd-c2fd-4bf0-979e-2fcf115d-068f-4a65-b443-134fb73e-fc99-41e4-92c4-12 2c5c6fe33fe0 1b4f96962bb7 6ecbada5574f 4ecc67cdeafd 20efa3c2-c498-4908-8af4-45a23a59-09cb-4906-8293a39dad1a-4b34-4f50-879e-0167b0c3-60fd-4fbd-b378-13 08bec70d2b36 f81c76781912 fba0ceac39ff 6014b72de249 2ba952cd-d5bf-4bb4-96fe-4fc4c56e-9e44-432a-9d1cf156fb67-e7e7-448c-b9c3a781da7a-1f22-4d1d-8975-14 0b5f0425d134 4219271a7844 02fd4e166d12 37165a13e6ca f851975d-1482-459b-b114c9cf4b47-8a85-488d-b33fc44107fb-70a4-4ff2-bedb-479bfe4c-7137-44dd-bb42-15 a4ab7900aa24 d1f921552ea0 48b9ecfd3d6d 5f73e87a9627 2b0bddcd-d73c-4ba6-b26e-026ffd7c-cff2-4daa-ae76d33a7b5d-15e5-4ba6-93ef-6e905a32-46a8-4555-8434-16 aaa785d50be7 6768d3283861 136bbcbf4946 a6b16580e873 5b9e2a68-7967-46e0-8e35-36117159-bb83-4b02-b1d0-950db98c-fb0f-4b78-bbeeb18c8c6e-0aa2-42ce-96d4-07f68fe03be6 bf9ae97faeca bc993a1f07e6 0e95f3f04ae5 fa60bc82-665e-4fe0-8f1fb377e5d5-563f-475d-8c6dc2237ca1-b7e3-40ab-b798eabecfd5-5ec5-4639-b32d-18 e1ea469301dc b9f85ad861fd b8a7675c2e2a e204b34af195 cfa8a834-54bd-4f47-99ca-76f02f30-28cd-4f15-88be-5158fc84-71e0-47a1-84e9aa4c2cd3-0f2c-4982-abd3-19 8912df32913b b3e446a391ae 9c64860d1fce be7b06facc87 In [14]: df3 = pd.merge(left=df2, right=df1, on="shipment_identifier", how="inner") df3.head() shipment_identifier origin_warehouse destination_store driver_identifier product on_time 449263b4-6c93-4f19-8b6abb75bf7d-c008-4267-bf92-5e9405de-a078-4b00-99c6c12025e6-6f9c-4728-8c3c-0 pants False 0d99a29fc637 6089cff5fe56 96564568b63c 9f840bde6f1a 449263b4-6c93-4f19-8b6abb75bf7d-c008-4267-bf92-5e9405de-a078-4b00-99c6c12025e6-6f9c-4728-8c3c-False pants 96564568b63c 0d99a29fc637 6089cff5fe56 9f840bde6f1a c12025e6-6f9c-4728-8c3c-449263b4-6c93-4f19-8b6abb75bf7d-c008-4267-bf92-5e9405de-a078-4b00-99c6-2 False pants 0d99a29fc637 6089cff5fe56 96564568b63c 9f840bde6f1a 449263b4-6c93-4f19-8b6abb75bf7d-c008-4267-bf92-5e9405de-a078-4b00-99c6c12025e6-6f9c-4728-8c3c-3 keyboards False 0d99a29fc637 6089cff5fe56 96564568b63c 9f840bde6f1a c12025e6-6f9c-4728-8c3cbb75bf7d-c008-4267-bf92-449263b4-6c93-4f19-8b6a-5e9405de-a078-4b00-99c6-4 keyboards False 6089cff5fe56 0d99a29fc637 96564568b63c 9f840bde6f1a df3.columns Index(['shipment_identifier', 'origin_warehouse', 'destination_store', 'driver_identifier', 'product', 'on_tim e'], dtype='object') df3.drop(['shipment_identifier','on_time'],axis=1, inplace=True) df3.head() destination_store driver_identifier origin_warehouse product **0** bb75bf7d-c008-4267-bf92-6089cff5fe56 pants 5e9405de-a078-4b00-99c6-96564568b63c c12025e6-6f9c-4728-8c3c-9f840bde6f1a bb75bf7d-c008-4267-bf92-6089cff5fe56 5e9405de-a078-4b00-99c6-96564568b63c c12025e6-6f9c-4728-8c3c-9f840bde6f1a pants bb75bf7d-c008-4267-bf92-6089cff5fe56 5e9405de-a078-4b00-99c6-96564568b63c c12025e6-6f9c-4728-8c3c-9f840bde6f1a pants bb75bf7d-c008-4267-bf92-6089cff5fe56 5e9405de-a078-4b00-99c6-96564568b63c c12025e6-6f9c-4728-8c3c-9f840bde6f1a **4** bb75bf7d-c008-4267-bf92-6089cff5fe56 5e9405de-a078-4b00-99c6-96564568b63c c12025e6-6f9c-4728-8c3c-9f840bde6f1a keyboards df3["quantity"] = np.random.random integers(1,101, size=len(df3)) df3 driver_identifier product quantity origin_warehouse destination_store bb75bf7d-c008-4267-bf92-5e9405de-a078-4b00-99c6c12025e6-6f9c-4728-8c3c-0 45 pants 96564568b63c 9f840bde6f1a 6089cff5fe56 c12025e6-6f9c-4728-8c3cbb75bf7d-c008-4267-bf92-5e9405de-a078-4b00-99c6-1 pants 48 96564568b63c 9f840bde6f1a 6089cff5fe56 bb75bf7d-c008-4267-bf92-5e9405de-a078-4b00-99c6c12025e6-6f9c-4728-8c3c-2 pants 65 96564568b63c 9f840bde6f1a 6089cff5fe56 c12025e6-6f9c-4728-8c3cbb75bf7d-c008-4267-bf92-5e9405de-a078-4b00-99c6-3 keyboards 68 6089cff5fe56 96564568b63c 9f840bde6f1a bb75bf7d-c008-4267-bf92-5e9405de-a078-4b00-99c6c12025e6-6f9c-4728-8c3ckeyboards 68 6089cff5fe56 96564568b63c 9f840bde6f1a b377e5d5-563f-475d-8c6deabecfd5-5ec5-4639-b32dfa60bc82-665e-4fe0-8f1f-b8a7675c2e2a 105 keyboards 65 e204b34af195 b9f85ad861fd 5158fc84-71e0-47a1-84e9aa4c2cd3-0f2c-4982-abd3animal 76f02f30-28cd-4f15-88be-9c64860d1fce 106 96 be7b06facc87 b3e446a391ae masks 5158fc84-71e0-47a1-84e9aa4c2cd3-0f2c-4982-abd3-107 70 76f02f30-28cd-4f15-88be-9c64860d1fce furniture b3e446a391ae be7b06facc87 aa4c2cd3-0f2c-4982-abd3-5158fc84-71e0-47a1-84e9-108 76f02f30-28cd-4f15-88be-9c64860d1fce 95 furniture b3e446a391ae be7b06facc87 5158fc84-71e0-47a1-84e9aa4c2cd3-0f2c-4982-abd3-109 76f02f30-28cd-4f15-88be-9c64860d1fce furniture 1 b3e446a391ae be7b06facc87 110 rows × 5 columns df3.columns = ["origin", "destination", "id", "product_id", "quantity"] df3.head(1) destination origin product_id quantity **0** bb75bf7d-c008-4267-bf92-6089cff5fe56 5e9405de-a078-4b00-99c6-96564568b63c c12025e6-6f9c-4728-8c3c-9f840bde6f1a pants df3 new = df3[["id", "product id", "quantity", "origin", "destination"]] In [24]: df3_new.head(1) Out[24]: id product_id quantity origin destination o c12025e6-6f9c-4728-8c3c-9f840bde6f1a 45 bb75bf7d-c008-4267-bf92-6089cff5fe56 5e9405de-a078-4b00-99c6-96564568b63c pants #df3_new.to_csv("shipment.csv", index=False) **Upload Data to Database** con = sq3.connect("shipment database.db") <sqlite3.Connection at 0x1d280607300> df0_new.to_sql("product", con, index=False) #Load to database df3_new.to_sql("shipment", con, index=False) #Load to database con.execute("Select * FROM sqlite_master").fetchall() Out[33]: [('table', 'product', 'product', 'CREATE TABLE "product" (\n"id" TEXT,\n "name" TEXT\n)'), table', 'shipment', 'shipment', 'CREATE TABLE "shipment" (\n"id" TEXT,\n "product id" TEXT,\n "quantity" INTEGER,\n "origin" TEXT,\n "des tination" TEXT\n)')] In [34]: pd.read_sql_query("Select * FROM sqlite_master", con) #Check all tables inside database Out[34]: name tbl_name rootpage type sql 2 CREATE TABLE "product" (\n"id" TEXT,\n "name"... 0 table product product 4 CREATE TABLE "shipment" (\n"id" TEXT,\n "prod... **1** table shipment shipment con.close() Python code done by Dennis Lam